Assignment 3

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Spring '24 - Anushree Kolhe
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In [1]: 1 import networkx as nx
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Section 1: Friendships Network

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In [2]:
         1 G = nx.read graphml('Data/highschool 2012.graphml')
            print(nx.info(G))
         4 | total_nodes = nx.number_of_nodes(G)
         5 total_edges = nx.number_of_edges(G)
        Graph with 180 nodes and 2220 edges
        /var/folders/_3/98b473cn1k993m8ln5gwyv6h0000gn/T/ipykernel_58026/2221535143.py:2: DeprecationWarning: info is de
        precated and will be removed in version 3.0.
          print(nx.info(G))
In [3]:
         1 highest_degree = max(G.degree(), key=lambda x: x[1])
            print(f'Node with highest degree: {highest_degree[0]}')
         3 print(f'Highest degree: {highest_degree[1]}\n')
        Node with highest degree: 826
        Highest degree: 56
In [4]:
            largest_cl_coeff = max(nx.clustering(G).items(), key=lambda x: x[1])
            print(f'Node with largest clustering coefficient: {largest_cl_coeff[0]}')
         3 | print(f'Largest clustering coefficient: {largest_cl_coeff[1]}\n')
        Node with largest clustering coefficient: 647
        Largest clustering coefficient: 1.0
```

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In [5]: 1 average_cl_coeff = round(nx.average_clustering(G), 4)
2 print(f'Average clustering coefficient of the n/w: {average_cl_coeff}\n')
```

Average clustering coefficient of the n/w: 0.4752

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In [6]: 1 total_Mnodes = 0
2 total_Fnodes = 0

for node, attributes in G.nodes(data=True):
    if attributes['gender'] == 'M':
        total_Mnodes += 1
    relse:
        total_Fnodes += 1

proportion_male = round(total_Mnodes/total_nodes, 4)
proportion_female = round(total_Fnodes/total_nodes, 4)
```

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In [7]: 1 print(f'Proportion of the nodes in the graph are male: {proportion_male}')
2 print(f'Proportion of the nodes in the graph are female: {proportion_female}\n')
```

Proportion of the nodes in the graph are male: 0.7333 Proportion of the nodes in the graph are female: 0.2667

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In [8]:
          1 # Expected / Calculated values
            expected_edges_ = {'M-M': 0, 'F-F': 0, 'M-F': 0}
            expected_edges_['M-M'] = round(proportion_male*proportion_male*total_edges) #m^2
            expected_edges_['F-F'] = round(proportion_female*proportion_female*total_edges) #f^2
             expected_edges_['M-F'] = round(2*proportion_male*proportion_female*total_edges) #2mf
             # Actual values
         10
            edges_ = {'M-M': 0, 'F-F': 0, 'M-F': 0}
         11
         12
            for node1, node2, attr in G.edges(data=True):
                 if G.nodes[node1]['gender'] != G.nodes[node2]['gender']:
         13
         14
                     edges['M-F'] += 1
         15
                 elif G.nodes[node1]['gender'] == 'M':
         16
                     edges ['M-M'] += 1
         17
                 else:
         18
                     edges_{['F-F']} += 1
In [9]:
          1 print(f'Expected edge values:\n{expected_edges_}\n')
            print(f'Actual edge values:\n{edges_}\n')
          3 print(f'Expected sum: {sum(expected_edges_.values())}')
             print(f'Actual sum: {sum(edges_.values())}\n')
         Expected edge values:
         {'M-M': 1194, 'F-F': 158, 'M-F': 868}
         Actual edge values:
         {'M-M': 1276, 'F-F': 182, 'M-F': 762}
         Expected sum: 2220
         Actual sum: 2220
In [10]:
          1 print(f'Expected edges M-F: {expected_edges_["M-F"]}\n')
          2 print(f'Actual edges M-F: {edges ["M-F"]}\n')
         Expected edges M-F: 868
         Actual edges M-F: 762
```

Section 2: Club Membership Network

What is the mean number of organizational affiliations per person in the data set? What is the mean number of members per organization?

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In [15]: 1 sum_org = 0
2 for org in organizations :
3    sum_org += len(set(G.neighbors(org)))
4    mean_org = sum_org/len(organizations)
```

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Function to measure similarity
In [16]:
          1 def similarity_measure(G, node_1, node_2):
           3
                 if (node 1 in people and node 2 in people) or (node 1 in organizations and node 2 in organizations):
          4
                     neighbors_n1 = set(G.neighbors(node_1))
           5
                     neighbors_n2 = set(G.neighbors(node_2))
                     sim = len(neighbors_n1.intersection(neighbors_n2))/len(neighbors_n1.union(neighbors_n2))
           6
           7
                     return sim
           8
          9
                 else:
          10
                     raise ValueError("Bipartite Graph!\n\tThe nodes need to be from the same group.")
In [17]:
          1 # Will throw an error as the groups are different.
           2 | similarity_measure(G, 'o11', 'p13')
         ValueError
                                                    Traceback (most recent call last)
         Input In [17], in <cell line: 2>()
               1 # Will throw an error as the groups are different.
         ----> 2 similarity_measure(G, 'o11', 'p13')
         Input In [16], in similarity_measure(G, node_1, node_2)
                   return sim
               7
               9 else:
                     raise ValueError("Bipartite Graph!\n\tThe nodes need to be from the same group.")
           --> 10
         ValueError: Bipartite Graph!
                 The nodes need to be from the same group.
In [18]: 1 | similarity_measure(G, 'o11', 'o6')
Out[18]: 0.3
         Function to get find the pair of nodes with highest similarity value
In [19]:
            def get_highest_similarity(G, grouptype):
          3
                 if grouptype == 'o': group = organizations
                 else: group = people
           5
           6
                 max_sim = 0
          7
                 pairs = []
          8
                 for i in range(len(group)-1):
          9
                     sim = similarity_measure(G, group[i], group[i+1])
          10
                      if sim > max_sim :
          11
                         max_sim = sim
          12
                         pairs = [group[i], group[i+1]]
          13
                 return(pairs, max_sim)
In [20]: 1 | get_highest_similarity(G, 'o')
```

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Out[20]: (['o6', 'o15'], 0.3)
In [21]: 1 get_highest_similarity(G, 'p')
Out[21]: (['p22', 'p3'], 0.75)
In []: 1
```